

# Operational Risk Management: A Practical Approach and its Regulatory Implications

Federal Reserve Bank of Boston

November 2001

# Operational Risk: Is a change required?

## Command and Control

Branch Operating Manual  
Mandated Controls  
Audit  
Monitoring resolutions of audit findings

Operational Risk is managed through  
better controls and better audit process

Managing Operational Risk Based on  
doing more of the same only better

## Inspire and Lead

Guiding Principles  
Boundaries  
Structured Self Assessments of Risk  
Monitoring of OP Risk Levels

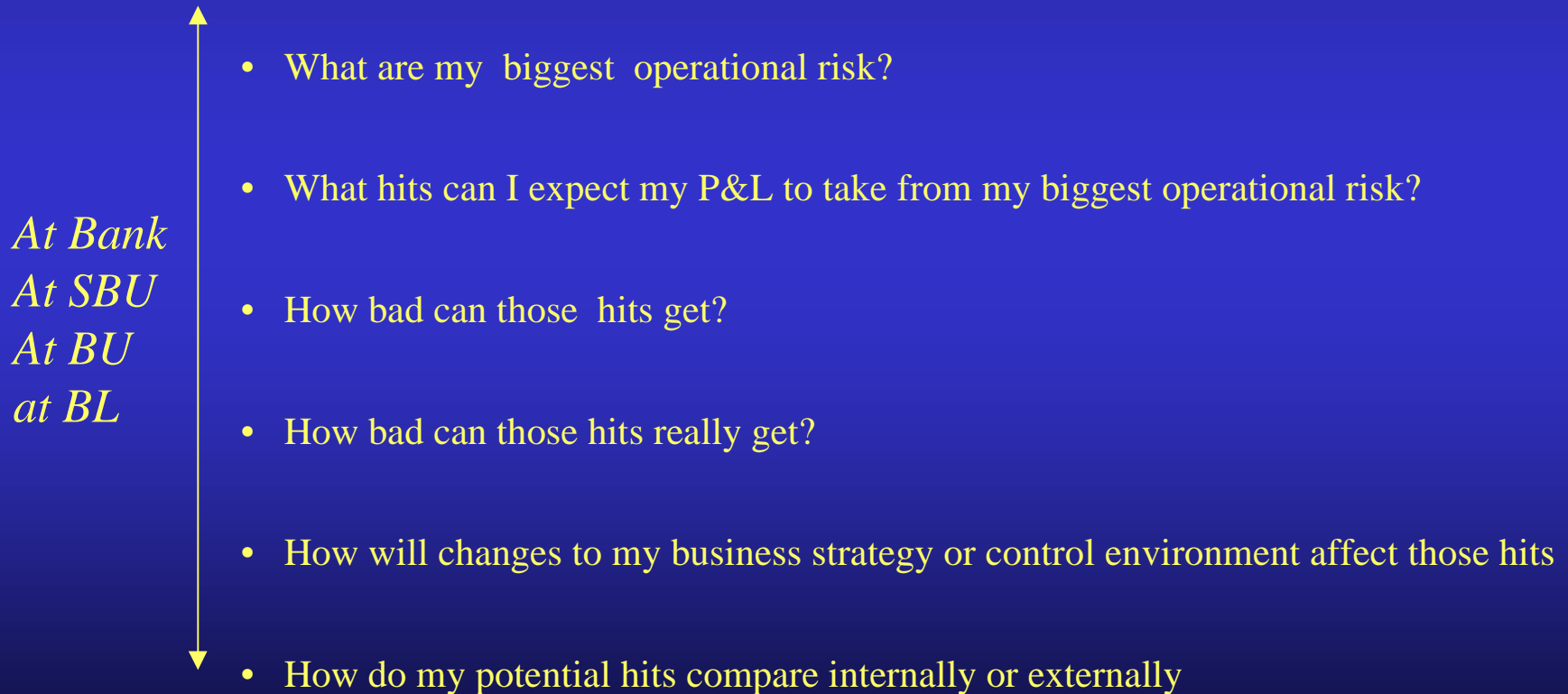
Operational Risk is managed through  
better risk identification and transparency  
of risk taken

A New Paradigm for Managing  
Operational Risk Based on \$ at Risk

# Under The New Paradigm

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**To Effective Manage Op Risk Business Leaders need to be able to know how much \$ are at Risk? More precisely answer these questions**



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How is operational risk measured

# Objective of Measuring Operational Risk

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- Provide an accurate view of the operational risk profile of the business over the next 12 months.
  - What is the expected losses from operational risk
  - What is the Worst Case Loss from operational risk
- Supports the analysis of Operational Risk
  - What are the top Op Risk
  - What is the Worst case loss under stress conditions
  - How will changes to my business strategy or control environment affect the potential.
  - How does the potential hit compare with other business units or other banks

# Measuring Operational Risk For better Management

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Based on analytic techniques widely used in the insurance industry to measure the financial impact of an operational failure

- **The foundation is**

- the historical operational loss experience
- deep understanding of what and why is at risk

- **The edifice is business judgement**

- similar to putting together a business plan
- judgement is used to supplement/ replace or enhance historical loss experience based inputs
- follows the same rigorous process as if all the inputs were historical loss data

- The measure is called OP VaR

- used for determining

- the expected loss from operational failures
  - the worst case loss at confidence level
  - the required economic and regulatory operational risk capital
  - concentration of operational risk
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# First Step

## Recognise Distinct Operational Risk Losses Types

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### **1. Legal Liability:**

includes client, employee and other third party lawsuits

### **2. Regulatory, Compliance and Taxation Penalties**

includes fines, or the cost of any other penalties, such as license revocations and associated costs - excludes lost / forgone revenue.

### **3. Loss of or Damage to Assets:**

reduction in value of the firm's non-financial asset and property

### **4. Client Restitution**

includes restitution payments (principal and/or interest) or other compensation to clients.

### **5. Theft, Fraud and Unauthorized Activities**

includes rogue trading

### **6. Transaction Processing Risk**

includes failed or late settlement, wrong amount or wrong counterparty

# For a line of business and loss type: The worst case loss (WCL) over the next 12 months

$$\begin{aligned}\text{WCL} &= \text{Expected Losses} \times \gamma \\ &= \text{Expected no of Losses} \times \text{Average Loss} \times \gamma\end{aligned}$$

**Expected no of losses**      the average number of legal liability, or transaction errors, or frauds etc over the next 12 months.

**Average Loss**      the average amount lost per legal liability, or per transaction error, or per frauds etc over the next 12 months

**$\gamma$**       Factor to convert the expected loss to worst case loss



# WCL Expressed in terms of Components of Expected Losses and Average Loss

$$\begin{aligned} \text{WCL} &= \text{Expected losses} \times \gamma \\ &= \text{Expected no of Losses} \times \text{Average Loss} \times \gamma \\ &= E_f \times PE_f \times E_s \times LGE \times \gamma \end{aligned}$$

$E_f$  = **Exposure for no of losses**  
eg no transactions, no of accounts, no of employees

$PE$  = **Expected Probability of an operational risk loss**  
eg Expected number of loss / the number of transactions

$E_s$  = **Exposure for loss amount**  
eg Avg transactions value, Avg accounts value, Avg employee compensation

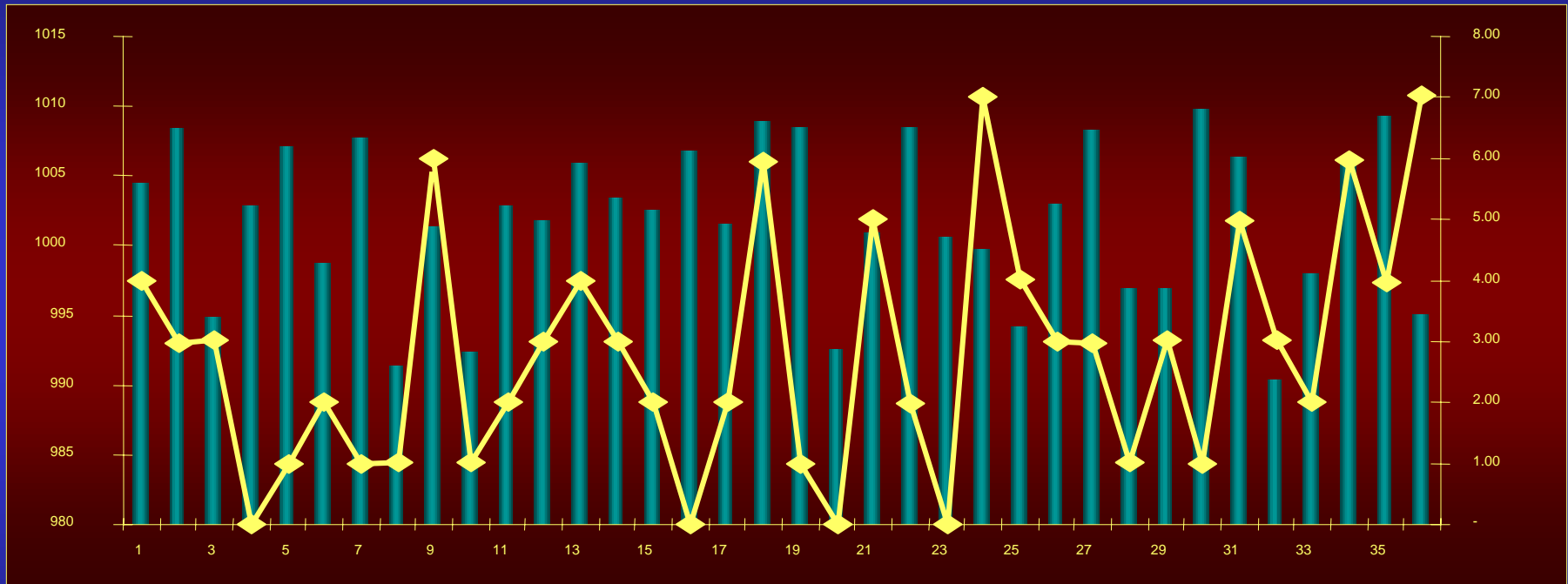
$LGE$  = **Average Loss Given Event Rate**  
eg average loss / Avg transactions value

$\gamma$  = **Factor to convert the expected loss to worst case loss**

# Op Risk Measurement Process

## Calculation of the Frequency ( PE)

Frequency (PE)



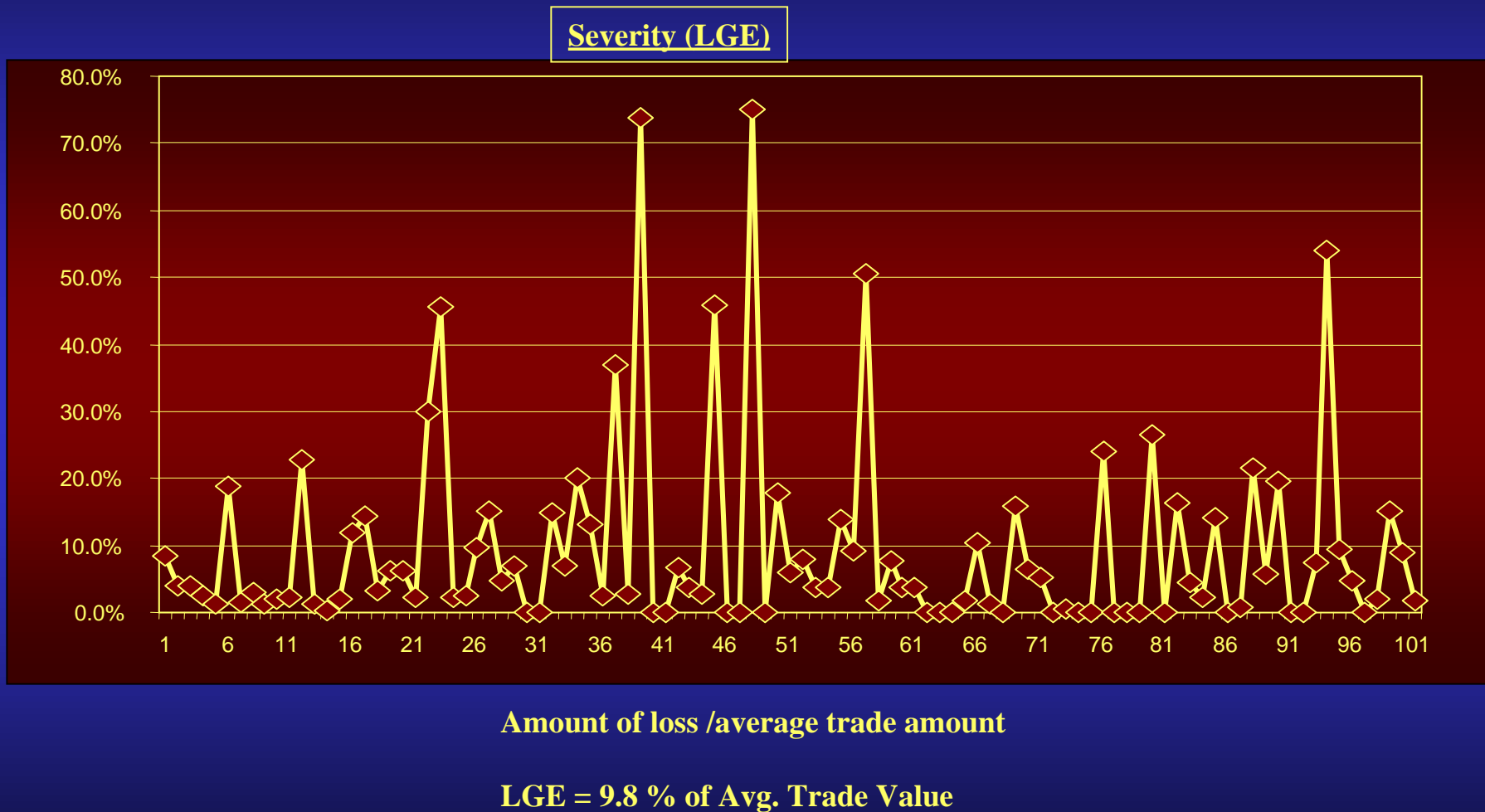
No of losses/ no of trades

PE = 2.8 per 10,000 Trades

can be desegregated for different type of trades  
or trade processing systems)

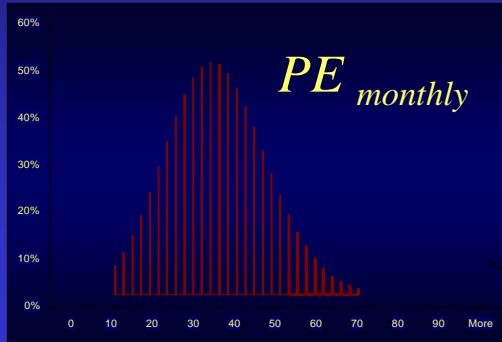
# Op Risk Measurement Process

## Calculation of the Severity (LGE)

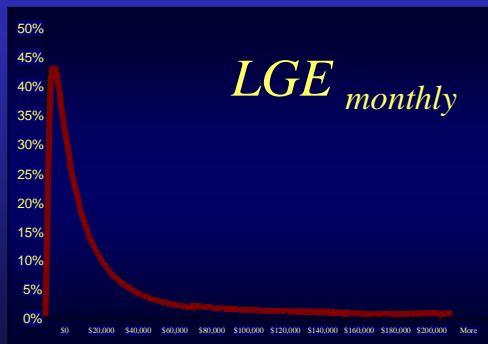
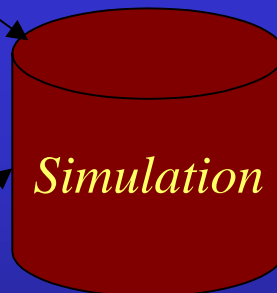


# Statistical Distributions and Simulate

## Function and Parameters

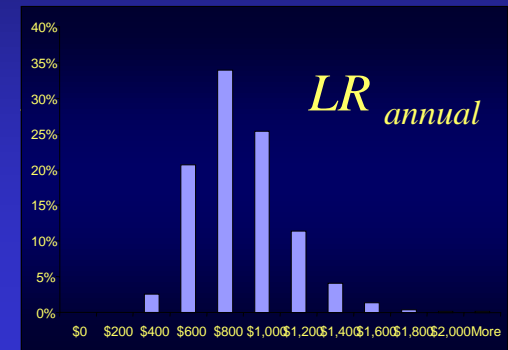
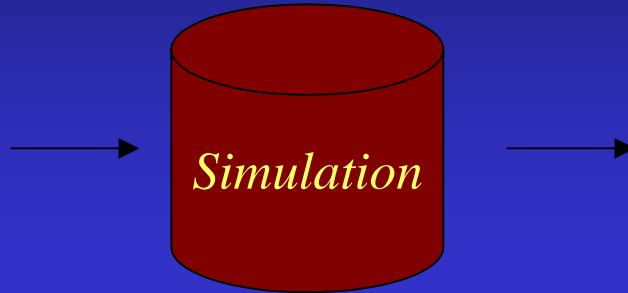


Function Poisson  
Mean PE 2.8 losses per  
10,000 transactions  
Std PE 2 events  
10,000 transactions



Function LogNormal  
Mean LGE 9.8 %  
Std PE 15%

# Annualize the Losses And Estimate Exposure



|              |     |
|--------------|-----|
| Av Loss Rate | 8%  |
| WCL          | 40% |
| Gamma        | 5   |

*With an Exposure of \$10mm the expected loss is \$.8mm ( \$10mm x 8%) and the worst case loss is \$4mm (\$10mm x 40%)*

# How Credible is the Result

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- Compare the PE and LGE derived from internal loss history with industry PE and LGE

## Example

- if external loss history shows one event per month
  - the internal loss history of 36 months is sufficient to determine with confidence the actual PE
- if external loss history shows one event in 10 years
  - the internal loss history of 36 months is not sufficient to determine with confidence the actual PE or the internally calculated PE is not credible
- When internal data is not credible, then the

$$\text{actual PE} = z_i \text{PE}_i + z_e \text{PE}_e$$

- Z are credibility factors and there are standard statistical methods for determining Z's

# Using external data

Insufficient internal loss data is supplemented with industry loss data

| Risk Types |  | TP |  |  |  |  |  |
|------------|--|----|--|--|--|--|--|
| LOB        |  |    |  |  |  |  |  |
| BU A       |  |    |  |  |  |  |  |
|            |  |    |  |  |  |  |  |
|            |  |    |  |  |  |  |  |
|            |  |    |  |  |  |  |  |

Loss rate/  
Exposure base  
\$ value of Transactions

Firm Specific Risk

TP

derived  
from  
internal  
data



General Industry Risk

TP

derived  
from  
industry  
data



$$\text{Capital} = \$\text{Value of Transactions} \times (\text{specific loss rate} + \text{general loss rate}) \times \gamma$$

$$\$4\text{mm} = \$10,000\text{m} \times \{ (Z)(8\%) + (1-Z)(12\%) \} \times 5$$

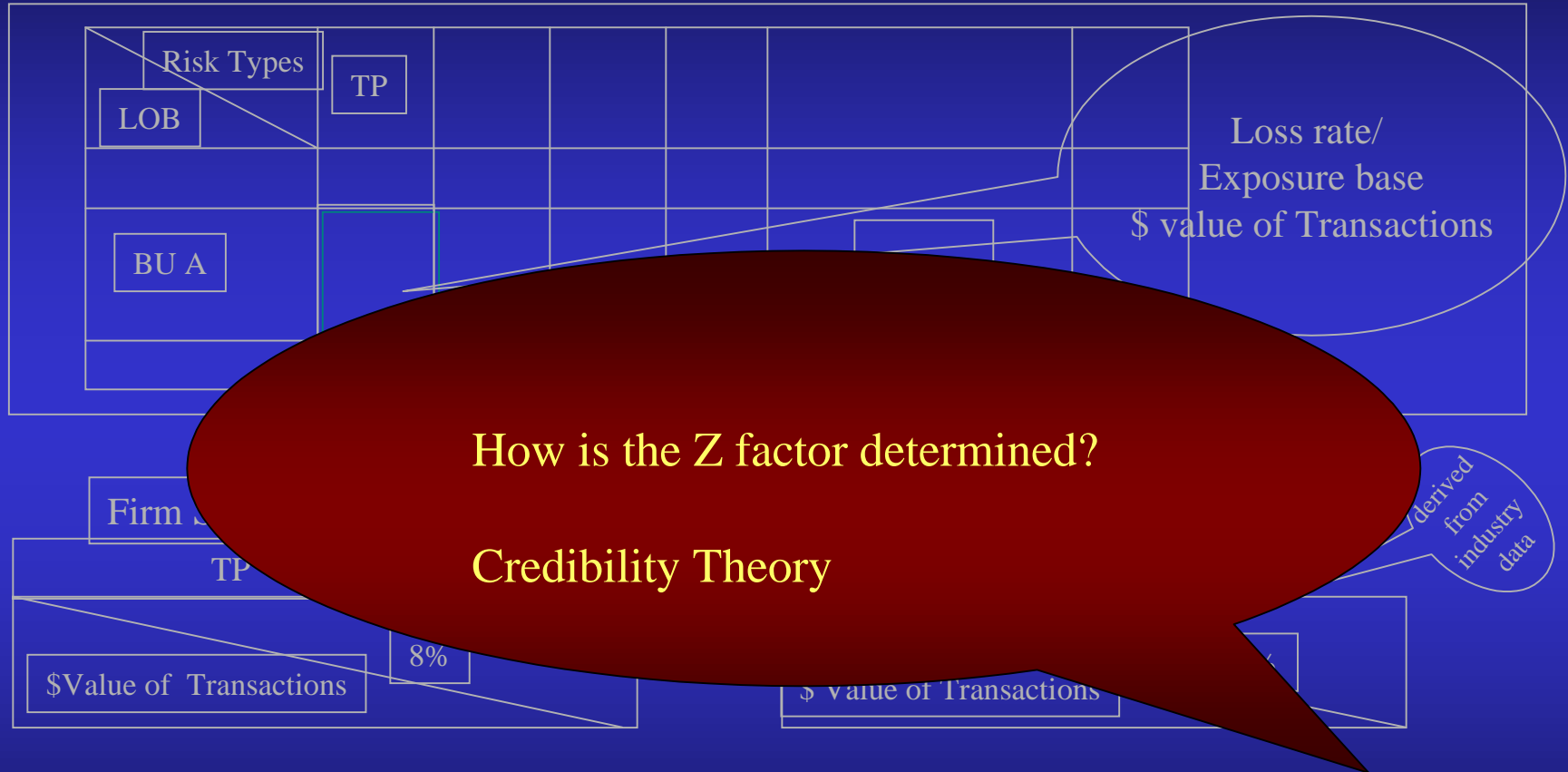
$$\$4.6\text{mm}$$

$$Z = 1$$

$$Z = .7$$

## Using external data

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$$Z = 1$$

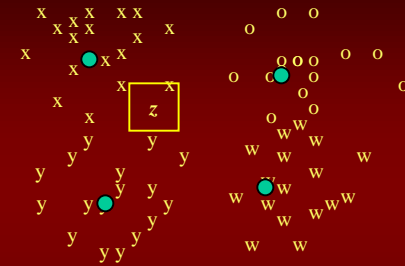
$$Z = .7$$



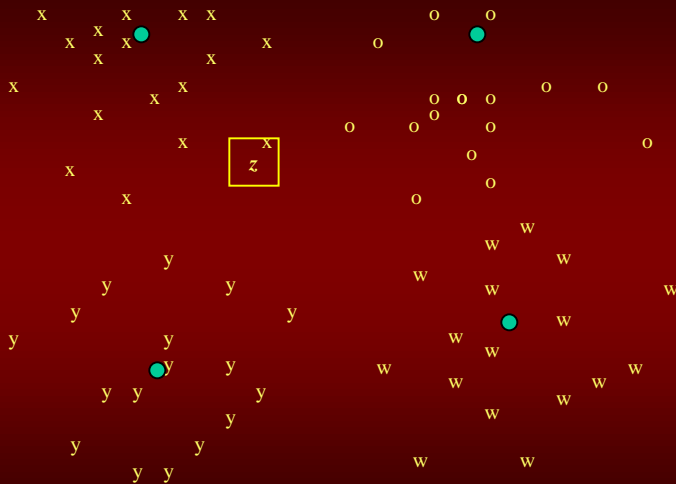
# Credibility



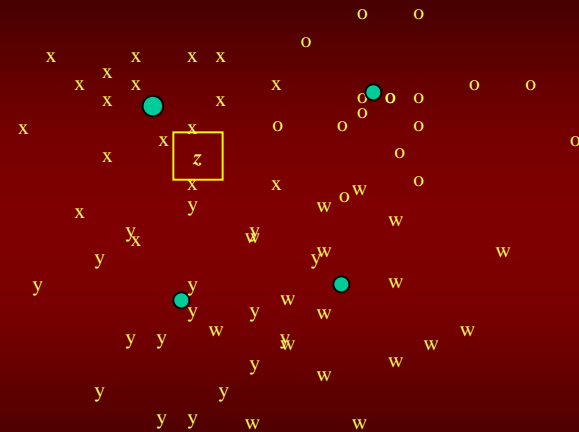
Means Are far Apart, Strong Clustering



Means Are Close, Strong Clustering



Means Are far Apart, Weak Clustering



Means Are Close, Weak Clustering

# How Credible is the Result

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## Example

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- ▶ When internal data is not credible, then the

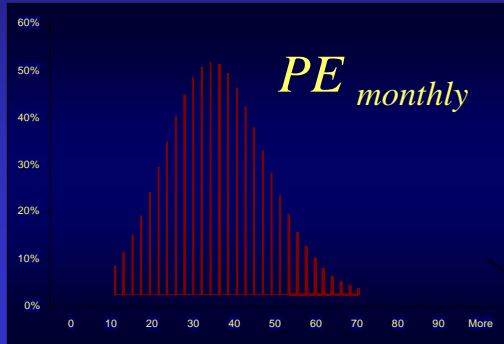
$$\text{Credible PE} = z_i \text{PE}_i + z_e \text{PE}_e$$

- ▶ When there are credibility factors and there are standard statistical methods for determining Z's

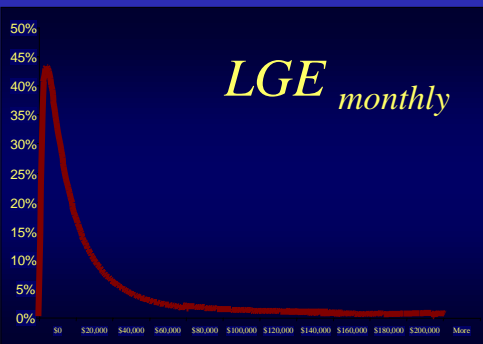
What happens when external data is insufficient

# Scenario Analysis

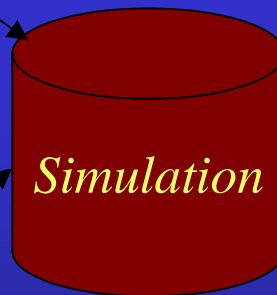
## Function and Parameters



|          |                                    |
|----------|------------------------------------|
| Function | Poisson                            |
| Mean PE  | 2.8 losses per 10,000 transactions |
| Std PE   | 2 events 10,000 transactions       |



|          |           |
|----------|-----------|
| Function | LogNormal |
| Mean LGE | 9.8 %     |
| Std PE   | 15%       |



These are estimated using  
Business and  
Risk Management  
Judgement



# Incorporating Scenario Analysis

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$$\text{actual PE} = z_i \text{PE}_i + z_e \text{PE}_e + z_s \text{PE}_s$$

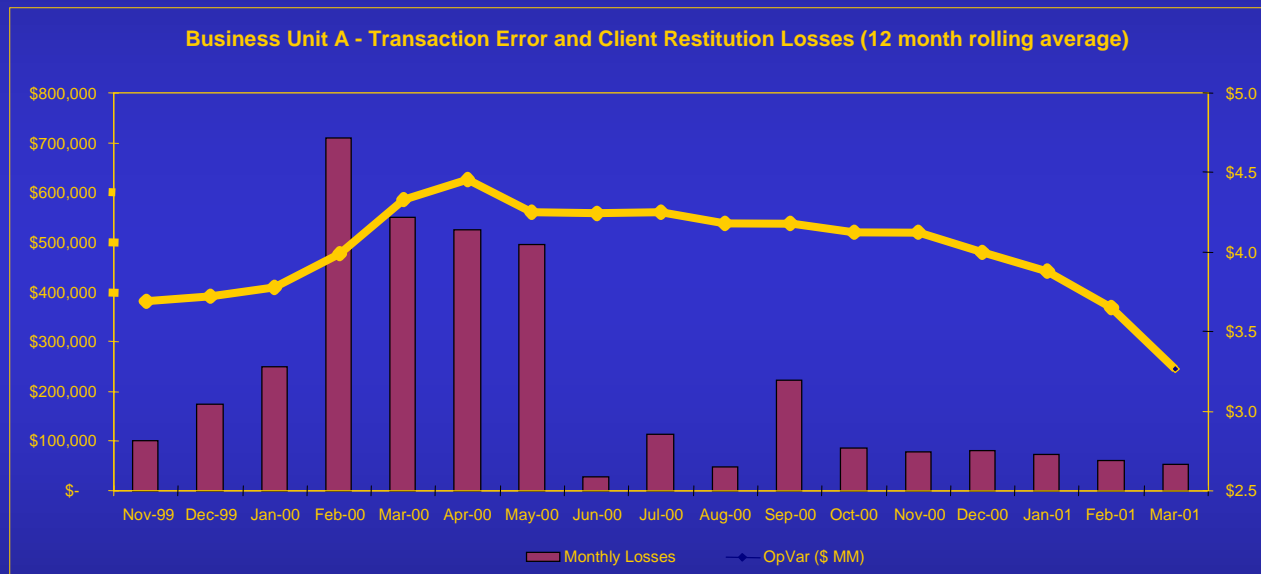


These are estimated using  
Statistics

These are estimated using  
Business and Risk Management  
Judgement

# Op VaR Reflects Changes in PE and LE over time

- Business Unit A



- Note the Lag
- How is  $\Delta$  BCE incorporated

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Use Qualitative Adjustments

# KRD's: Key Risk Drivers

- Used to monitor changes operational risk for each business and for each loss type **before** the change in loss experience can be observed ( ie lag and low frequency events)
- Incorporated into Op VaR, by modifying the risk determined by loss history and can be used to reward and punish for positive or negative changes in risk profile
- Objective standard measure eg a standard score
- Needs to be developed
- Can be as simple as the audit score or as sophisticated as the 100 metrics used by some banks ( eg % of book daily independently reevaluated, % of system down time, age of systems)

## Example of How KRD can be used to Adjust Op VaR

|                      | $\Delta$ KRD % | $\Delta$ Op VaR % |
|----------------------|----------------|-------------------|
|                      | +20            | -15               |
|                      | +10            | -10               |
| Op VaR <sub>LE</sub> | 0              | 0                 |
|                      | -10            | +15               |
|                      | -20            | +25               |

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Is there an alternative to the scorecard approach to the Qualitative Adjustment or more precisely to incorporating the  $\Delta$  BCE



# Incorporating Scenario Analysis

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$$\text{actual PE} = z_i \text{PE}_i + z_e \text{PE}_e + z_s \text{PE}_s$$



These are estimated using Statistics

These are estimated using Business and Risk Management Judgement

Use the scenario involving the  $\Delta$  BCE

- Business and Risk Management must estimate the effect of the  $\Delta$  BCE on PE, LGE and  $\gamma$

# General OP VaR Methodology

$$\begin{aligned} \text{WCL} &= \text{Expected no of Losses} \times \text{Average Loss} \times \gamma \\ &= E_f \times E_s \times \text{PE} \times \text{LGE} \times \gamma \end{aligned}$$

$$E_f \times E_s = E = (1 - Z_{le}) E_h + Z_{le} E_{le}$$

$E_h$  = 12 month average exposure  
 $E_{LE}$  = latest estimate from BM Judgement  
 $Z_{le}$  is from RM and BM Judgement

$$\text{PE} = Z_h \text{PE}_h + Z_e \text{PE}_e + Z_{bce} \text{PE}_{bce}$$

$\text{PE}_h$  = 36 month average rate from internal loss experience  
 $\text{PE}_e$  = 36 month average rate from external loss experience  
 $\text{PE}_{bce}$  = Scenario analysis (RM and BM Judgement)

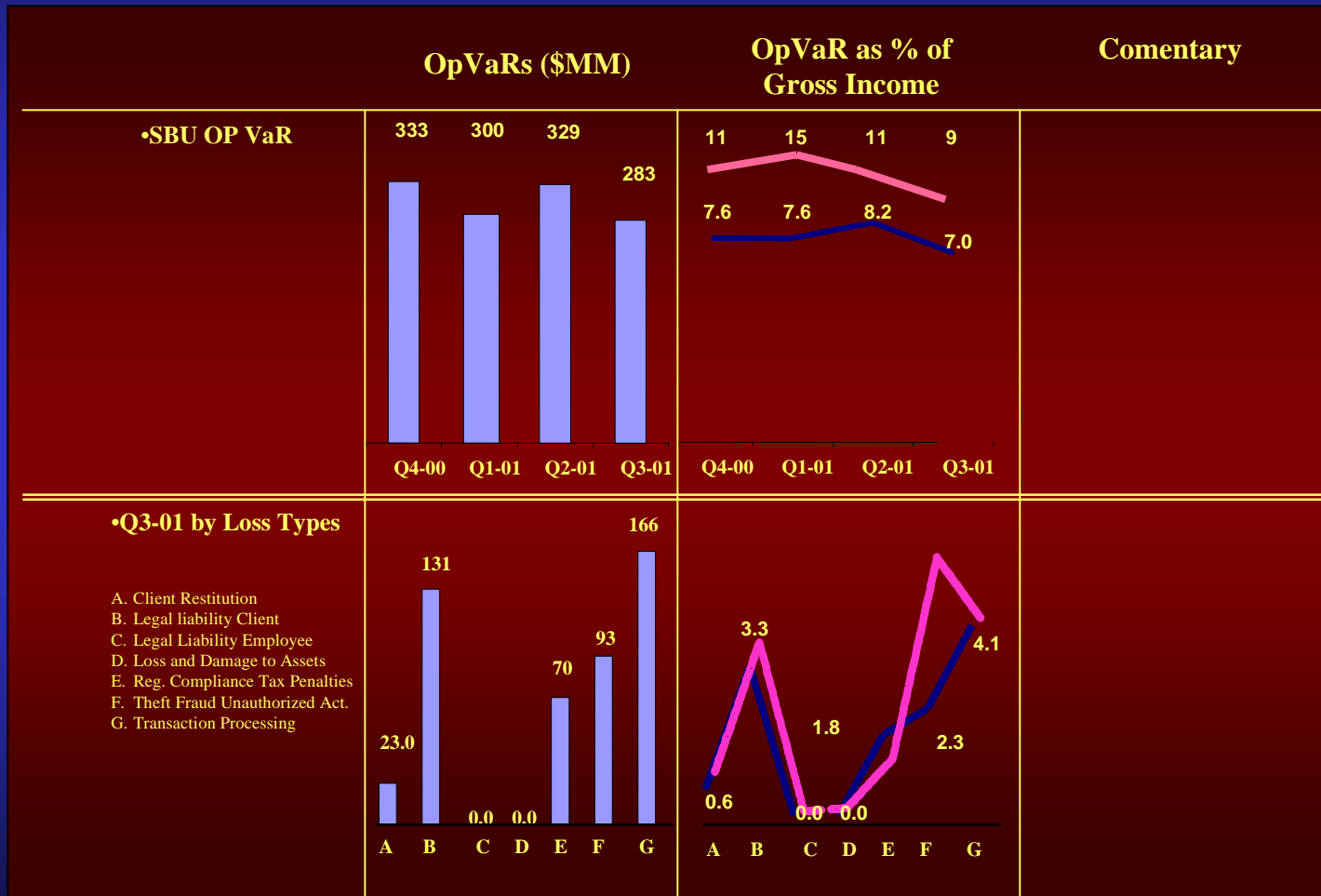
$Z_h$  and  $Z_e$  Calculated using statistical credibility theory  
 $Z_{bce}$  is from RM and BM Judgement

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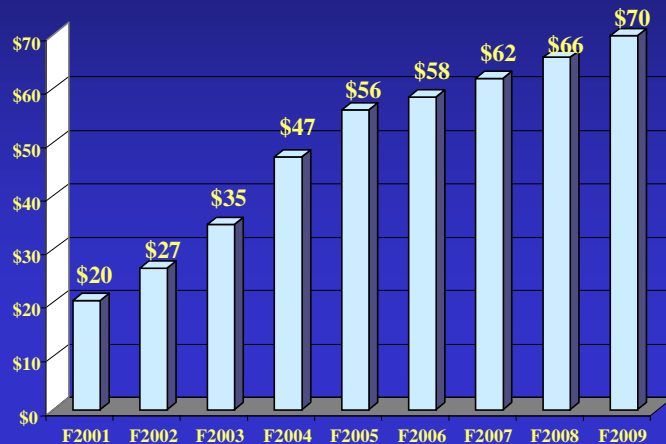
# The *AMOR* Report

Analysis and  
*M*onitoring of  
Operational  
*R*isk

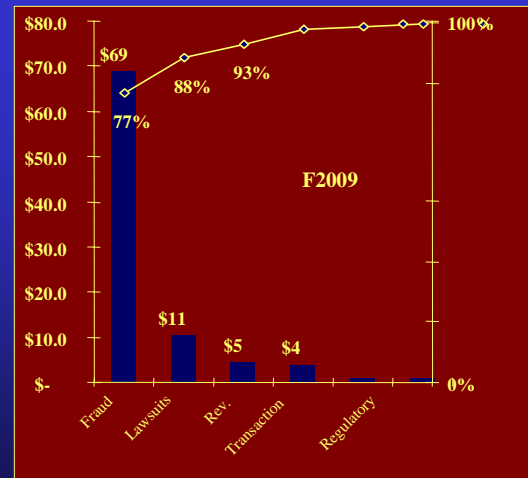
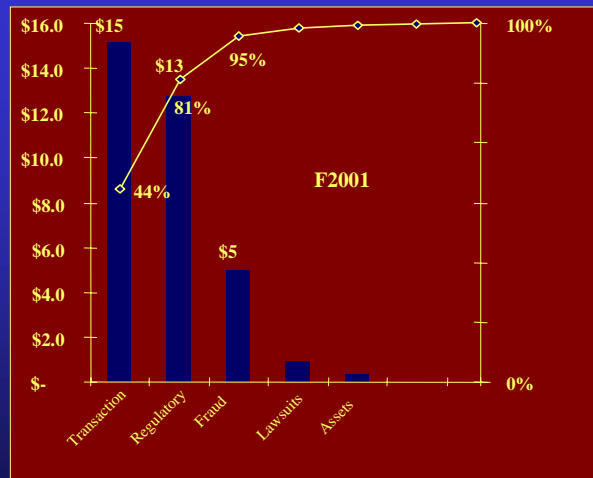
# AMOR



# Start up Op VaR



- OP VaR increases 2.5 times over 9 years compared to account balance growth 208 times:
  - Reduction in infrastructure build.
  - Reduction in fraud rates because of business maturing..
- Composition of Op VaR changes over the 9 year horizon
  - First year, 81% of risk is Transaction risk reflecting infrastructure (kiosks) build up
  - Ninth year, 77% of the risk is Theft and Fraud.
  - Historical Proxy losses rates have been used





# Decomposing Expected No Of Losses

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Expected no of Losses can be decomposed into the a measure of the amount of the business activity that can gives rise to the loss and the propensity for losses given that activity.

- ▶ This allows comparison of operational risk over time, by separating out
  - how much of the change is due to the change in the amount of business activity and
  - how much is due to a change in the propensity for losses
  
- ▶ the measure of the amount of business activity should correlate with the number op expected operational risk losses, this measure is usually referred to as the frequency exposure and denoted as  $E_f$ 
  - For example: in transaction risk,  $E_f$  may be total number of transaction processed
  
- ▶ the propensity for loss is the probability that business activity gives rise to a loss and is denoted by PE

$$\text{Expected no of Losses} = E_f \times PE$$

# Decomposing Average Loss

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Average loss can be decomposed into the average of amount at risk per loss event and the percentage lost per loss event

- ▶ this allows the comparison of operational risk over time, by separating out
  - how much of the change is due to the change in the amount at risk per loss event and
  - how much is due to a change in the percentage lost per loss event
  - This decomposition is especially useful for risk management, when action ca
- ▶ the measure of the amount at risk should correlate with average loss per loss event, this measure is usually referred to as the severity exposure and denoted as  $E_s$ 
  - For example: in transaction risk,  $E_s$  may be average value of transaction processed
- ▶ the percentage lost of the amount at risk per loss event is denoted by LGE ( loss given event)

$$\text{Average Loss} = E_s \times \text{LGE}$$